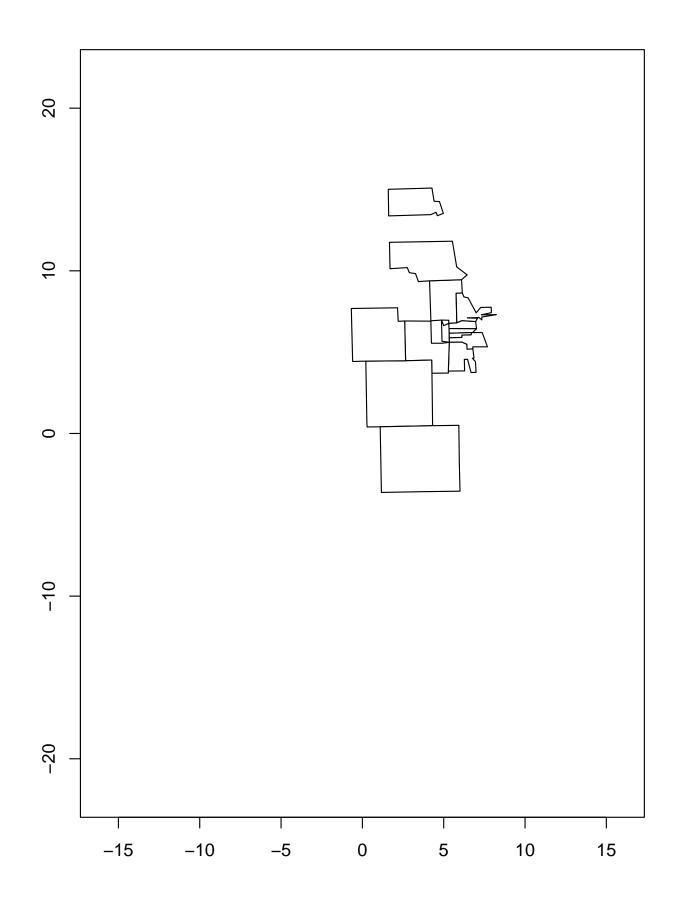
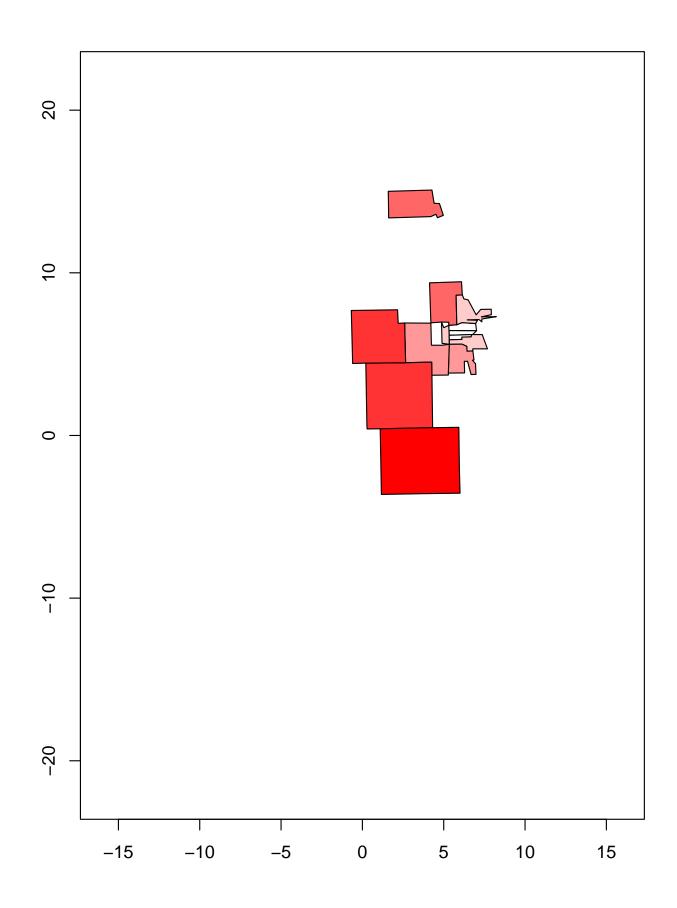
Mod …'n __ c …of A' out 'on on Ad' A n it co 'n Cni€ o

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Crite o A real M die id Die is people on Medicaid in Chicago Medicaid (IDPA) claims data





Crite o A' out on De s st

Ozone: 11 monitors

- Hourly ozone
- Max-8-hr average ozone
- Ordinary spatial Kriging assuming isotropy
- Matern covariance function
- Predicted values of ozone (and se) at each
 ZIP centroid
- Spatially resolved exposure measure!

PM10: 18 monitors

- **Observations spaced 6 days apart**
- Monitor and day e ect ANOVA model to predict PM10 for any given monitor on any given day
- Daily average based on the tted values

Cric o Cric nd o. n D Cric Weather: Average of O'Hare and Midway:

-

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Why use BA (Albuterol)? { often-used non-steroid bronchodilators { speci c to asthma { running prescriptions, quick use { large counts

Q: Is using BA prescriptions as a marker for asthma outcome reasonable?

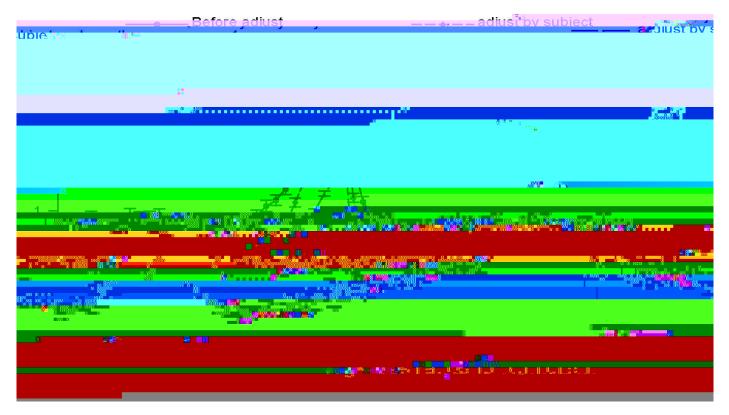
 Examine association between BA and hospital admits / ED visits (more traditional outcomes), d by one or more days

Odds ratio (OR)

{ crude OR

{ subject-adjusted OR

{ subject- and time-adjusted OR



(Note: positive lag BA before ED)

Problems with crude OR:

... Artifactually

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- Description observed period (over space and time) disease risk within ZIP code (≥ ... n⁻)
- p d c d disease risk (given data) po n y po n

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description of disease incidence

```
... spatial interpolation
```

! disease risk pp'n

place spatial epidemiologic studies in context by displaying background risk

•_

interpret (variation in) disease risk at po'n versuse versuse level

- smooth variation across areal units
- account for (di erences in) sampling
 variability across areal units
- borrow information across units

understand areal data aggregation

improve statistical e ciency in spatial regression models

App o c Mod A Sion
R is o R Mod A Sion
Then ...

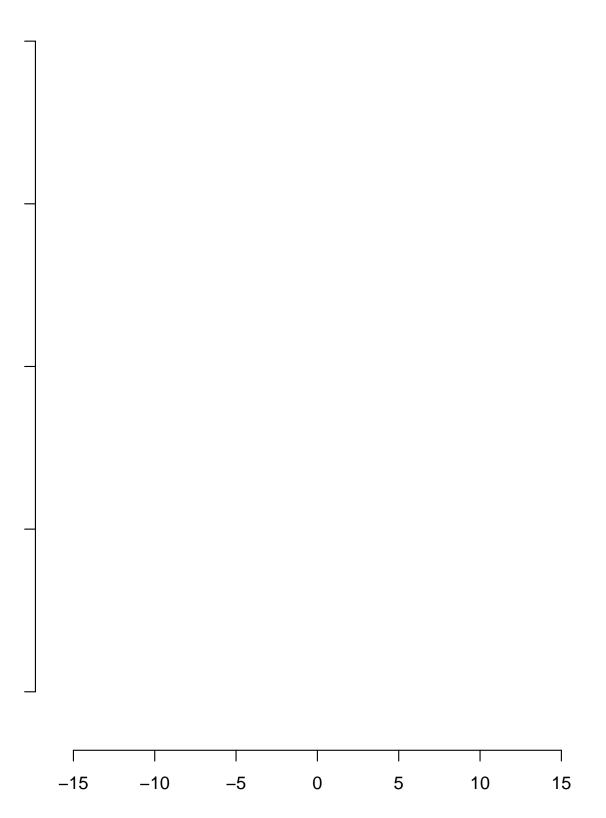
 Y_{i} Poisson (person-time $e^{\overline{R}_{i}}$) where R_{i} avef R g \sim o \sim over ZIP i

Converts point \mathfrak{E} problem in R to \mathfrak{E} \mathfrak{E} problem in R_i

An estimate of R_i is

$$\widehat{R}_{i}$$
 $\left(\frac{\text{\# events}}{\text{person-time}}\right)$

the observed log risk in ZIP i \hat{R}_i 's are used to p d'c R at every



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> new statistical space-time models exploit physical models such as CMAQ improve air pollution exposure measurement

- 2. Validation of Beta-agonist as Asthma Outcome
- 3. Mapping asthma outcomes aggregated by ZIP-code
- 4. A R d R · · R A y · f
- 5. p. topon the dimension of the second seco

C _ zon A AS Co n S o

Vanja Dukic (Health Studies)

Dana Draghicescu (CISES, Research Associate) John Frederick (Geophysical Sciences) Edward Naureckas (Medicine) Alexis Zubrow (CISES, Programmer/Analyst) Xiaoming Bao (Statistics, MS Student) Lon 'u' d'n An y' of A and u' co De t

szd od.

$$Y_{ijt} \qquad e^{\mathbf{R}(\mathbf{x}_{i}) + \mathbf{S}_{i} + \mathbf{z}_{i} t}$$

where:

- R is baseline or R at at
- S_{ijk} all unobserved factors for person
- i j in time \window" k
- z_{ijt} covariates for person i j, day t
- - adjusted for spatially and within-person slowly-varying factors
 - has improved statistical e ciency

M · zrma

Lon 'v' d'n Di An y' .: 'n y Mod .: Adv.

Odds ratios for Risk of BA Prescription Fill Conditional Logit Model

	Odds ratio	St. Err.	Z
ozone (20ppb)	1.005	.007	0.81
pm10 (15mg/m3)	.999	.004	-0.22
log(pollen) (std)	1.009	.004	2.11
temp (5F)	.984	.003	-5.20
rel hum (10%)	.997	.003	-0.95

Average Odds Ratios across 51 ZIP Codes				
	Odds ratio	St. Err.	Z	
ozone (20ppb)	1.032	.022	1.47	